

```
In [ ]: import pandas as pd
import numpy as np
```

```
In [ ]: df = pd.read_csv('titanic.csv')
df
```

Out[]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked	
	0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
	1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	0	PC 17599	71.2833	C85	C
	2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
	3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
	4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S

	886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.0000	NaN	S
	887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.0000	B42	S
	888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.4500	NaN	S
	889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.0000	C148	C
	890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.7500	NaN	Q

891 rows × 12 columns

```
In [ ]: df.describe()
```

Out []:

	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare
count	891.000000	891.000000	891.000000	714.000000	891.000000	891.000000	891.000000
mean	446.000000	0.383838	2.308642	29.699118	0.523008	0.381594	32.204208
std	257.353842	0.486592	0.836071	14.526497	1.102743	0.806057	49.693429
min	1.000000	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000
25%	223.500000	0.000000	2.000000	20.125000	0.000000	0.000000	7.910400
50%	446.000000	0.000000	3.000000	28.000000	0.000000	0.000000	14.454200
75%	668.500000	1.000000	3.000000	38.000000	1.000000	0.000000	31.000000
max	891.000000	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200

```
In [ ]: df.nunique()
```

Out []:

```
PassengerId    891
Survived        2
Pclass          3
Name            891
Sex              2
Age             88
SibSp           7
Parch           7
Ticket         681
Fare           248
Cabin          147
Embarked        3
dtype: int64

Checking for Duplicate rows
```

```
In [ ]: df.duplicated().sum()
```

Out []: 0

Checking for Null values

```
In [ ]: df.isna().sum()
```

```
Out[ ]: PassengerId      0
Survived                0
Pclass                 0
Name                   0
Sex                   0
Age                  177
SibSp                 0
Parch                 0
Ticket                0
Fare                  0
Cabin                687
Embarked              2
dtype: int64
```

Cabin column is dropped because of too many missing values Name and Ticket dropped because it is unique at each column

```
In [ ]: df2 = df.drop(columns=['Name', 'Ticket', 'Cabin'])
```

Dropped 2 rows where embarked is null

```
In [ ]: df2.dropna(inplace=True, subset=['Embarked'])
```

```
In [ ]: data = df2.values
X = data[:, 2:]
print(X)
```

```
[[3 'male' 22.0 ... 0 7.25 'S']
[1 'female' 38.0 ... 0 71.2833 'C']
[3 'female' 26.0 ... 0 7.925 'S']
...
[3 'female' nan ... 2 23.45 'S']
[1 'male' 26.0 ... 0 30.0 'C']
[3 'male' 32.0 ... 0 7.75 'Q']]
```

```
In [ ]: y = data[:,1]
print(y)
```

```
[0 1 1 1 0 0 0 0 1 1 1 1 0 0 0 1 0 1 0 1 0 1 1 1 0 1 0 0 1 0 0 1 1 0 0 0 1
0 0 1 0 0 0 1 1 0 0 1 0 0 0 0 1 1 0 1 1 0 1 0 0 0 0 0 1 1 0 1 0 0 0 0 0 1
0 0 0 1 1 0 1 1 0 1 1 0 0 1 0 0 0 0 0 0 0 0 0 1 1 0 0 0 0 0 0 1 1 0 1 0 0
0 0 0 0 0 0 0 0 0 0 0 1 0 1 0 1 1 0 0 0 0 1 0 0 0 0 1 1 0 0 0 1 0 0
0 0 1 0 0 0 0 1 0 0 0 0 1 0 0 0 1 1 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 1 1 0
1 1 0 0 1 0 1 1 1 1 0 0 1 0 0 0 0 0 1 0 0 1 1 1 0 1 0 0 0 1 1 0 1 0 1 0 0
0 1 0 1 0 0 0 1 0 0 1 0 0 0 1 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 0 0 1 1 1 1
0 1 0 0 0 0 0 1 1 1 0 1 1 0 1 1 0 0 0 1 0 0 0 1 0 0 1 0 1 1 1 0 0 0 0 0
0 1 1 1 1 0 1 0 1 1 1 0 1 1 1 0 0 0 1 1 0 1 1 0 0 1 1 0 1 0 1 1 1 0 0 0
1 0 0 1 1 0 1 1 0 0 0 1 1 1 1 0 0 0 0 0 0 1 0 1 1 0 0 0 0 0 0 1 1 1 1 1
0 0 0 0 1 1 0 0 0 1 1 0 0 0 1 0 1 1 1 0 1 1 0 0 0 0 1 1 0 0 0 0 0 1
0 0 0 0 1 0 1 0 1 1 0 0 0 0 0 0 0 0 1 1 0 1 1 1 0 0 1 0 1 0 0 1 0 0 1 1
1 1 1 1 1 0 0 0 1 0 1 0 1 1 0 1 0 0 0 0 0 0 0 0 1 0 0 1 1 0 0 0 0 0 1 0 0
0 1 1 0 1 0 0 1 0 0 0 0 0 0 1 0 0 0 0 0 0 0 1 0 1 1 0 1 1 0 1 1 0 0 1 0 1
0 1 0 0 1 0 0 1 0 0 0 1 0 0 1 0 1 0 1 0 1 1 0 0 1 0 0 1 1 0 1 1 0 0 1 1 0
1 0 1 1 0 0 0 0 0 0 0 0 0 1 1 1 1 0 0 1 1 0 1 1 1 0 0 0 1 0 1 0 0 0 1 0
0 0 0 1 0 0 1 1 0 0 0 1 0 0 1 1 1 0 0 1 0 0 1 0 0 1 0 0 1 1 0 0 0 0 1 0 0
1 0 1 0 0 1 0 0 0 0 0 1 0 1 1 1 0 1 0 1 0 1 0 1 0 0 0 0 0 0 1 0 0 0 1 0 0
0 0 1 1 0 0 1 0 0 0 1 0 1 0 1 0 0 0 0 0 0 0 1 1 1 1 0 0 0 0 1 0 0 1 1 0 0
0 0 1 1 1 1 0 1 0 0 0 1 1 0 0 1 0 0 0 1 0 1 1 0 0 1 0 0 0 0 0 0 1 0 0 1
0 1 0 1 0 0 1 0 0 1 1 0 0 1 1 0 0 0 1 0 0 1 1 0 1 0 0 0 0 0 0 0 0 1 0 0 1
0 1 1 1 0 0 0 0 1 0 1 0 0 0 0 0 0 0 1 1 0 0 0 1 1 1 0 0 0 0 1 0 0 0 0 0
0 0 0 0 0 1 1 0 1 0 0 0 1 1 1 1 0 0 0 1 0 0 1 1 0 0 1 0 0 0 0 0 0 1 0 0 0
1 0 1 1 1 1 0 0 0 1 0 0 1 1 0 0 1 0 1 0 0 1 1 0 0 0 1 1 0 0 0 0 0 0 1 0 1
0]
```

```
In [ ]: from sklearn.model_selection import train_test_split
```

Splitting the array into test and train sets

```
In [ ]: X_train, X_test, y_train, y_test = train_test_split(X,y, test_size=0.2, random_state=1)
```

```
In [ ]: print(X_train.shape)
print(X_test.shape)
print(y_train.shape)
print(y_test.shape)
```

```
(711, 7)
(178, 7)
(711,)
(178,)
```

```
In [ ]: print(X_train)
```

```
[[3 'male' nan ... 0 8.05 'S']
 [3 'male' 19.0 ... 0 10.1708 'S']
 [3 'male' nan ... 0 7.75 'Q']
 ...
 [3 'male' 26.0 ... 0 14.4542 'C']
 [2 'male' 44.0 ... 0 26.0 'S']
 [3 'male' 21.0 ... 0 8.05 'S']]
```

```
In [ ]: print(X_test)
```

```
[[2 'female' 36.0 ... 0 13.0 'S']
 [2 'female' 50.0 ... 1 26.0 'S']
 [3 'male' 48.0 ... 0 7.8542 'S']
 ...
 [1 'male' 22.0 ... 0 135.6333 'C']
 [3 'male' 22.0 ... 0 8.05 'S']
 [2 'male' 24.0 ... 0 10.5 'S']]
```

```
In [ ]: from sklearn.impute import SimpleImputer
```

Imputer to fill null values using mode

```
In [ ]: imputer = SimpleImputer(missing_values=np.nan, strategy='most_frequent')
```

Replacing null values in age with mode

```
In [ ]: imputer.fit(X_train[:, 2:3])
```

```
Out[ ]: SimpleImputer
SimpleImputer(strategy='most_frequent')
```

```
In [ ]: X_train[:, 2:3] = imputer.transform(X_train[:, 2:3])
print(X_train)
```

```
[[3 'male' 24.0 ... 0 8.05 'S']
 [3 'male' 19.0 ... 0 10.1708 'S']
 [3 'male' 24.0 ... 0 7.75 'Q']
 ...
 [3 'male' 26.0 ... 0 14.4542 'C']
 [2 'male' 44.0 ... 0 26.0 'S']
 [3 'male' 21.0 ... 0 8.05 'S']]
```

```
In [ ]: X_test[:, 2:3] = imputer.transform(X_test[:, 2:3])
print(X_test)
```

```
[[2 'female' 36.0 ... 0 13.0 'S']
 [2 'female' 50.0 ... 1 26.0 'S']
 [3 'male' 48.0 ... 0 7.8542 'S']
 ...
 [1 'male' 22.0 ... 0 135.6333 'C']
 [3 'male' 22.0 ... 0 8.05 'S']
 [2 'male' 24.0 ... 0 10.5 'S']]
```

Min Max Feature Scaling on columns Age and Fare

```
In [ ]: from sklearn.preprocessing import MinMaxScaler
min_max_scaler = MinMaxScaler(feature_range=(0,1))
```

```
In [ ]: X_train[:,(2,5)] = min_max_scaler.fit_transform(X_train[:,(2,5)])
X_test[:,(2,5)] = min_max_scaler.transform(X_test[:,(2,5)])
```

```
In [ ]: print(X_train)
```

```
[[3 'male' 0.2963056044232219 ... 0 0.03060836501901141 'S']
 [3 'male' 0.2334757476752953 ... 0 0.038672243346007606 'S']
 [3 'male' 0.2963056044232219 ... 0 0.029467680608365018 'Q']
 ...
 [3 'male' 0.32143754712239253 ... 0 0.05495893536121673 'C']
 [2 'male' 0.5476250314149284 ... 0 0.09885931558935361 'S']
 [3 'male' 0.2586076903744659 ... 0 0.03060836501901141 'S']]
```

```
In [ ]: print(X_test)
```

```

[[2 'female' 0.4470972606182458 ... 0 0.049429657794676805 'S']]
[[2 'female' 0.6230208595124404 ... 1 0.09885931558935361 'S']]
[[3 'male' 0.5978889168132697 ... 0 0.029863878326996197 'S']]
...
[[1 'male' 0.2711736617240512 ... 0 0.515715969581749 'C']]
[[3 'male' 0.2711736617240512 ... 0 0.03060836501901141 'S']]
[[2 'male' 0.2963056044232219 ... 0 0.039923954372623575 'S']]

```

```

In [ ]: from sklearn.compose import ColumnTransformer
        from sklearn.preprocessing import OneHotEncoder

```

Encoding Sex and Embarked columns using OneHotEncoder

```

In [ ]: ct = ColumnTransformer(transformers=[('encode', OneHotEncoder(),[1,-1])], remainder='passthrough')

```

```

In [ ]: X_train = ct.fit_transform(X_train)
        X_test = ct.transform(X_test)

```

```

In [ ]: print(X_train)

```

```

[[0.0 1.0 0.0 ... 0 0 0.03060836501901141]
 [0.0 1.0 0.0 ... 0 0 0.038672243346007606]
 [0.0 1.0 0.0 ... 0 0 0.029467680608365018]
 ...
 [0.0 1.0 1.0 ... 1 0 0.05495893536121673]
 [0.0 1.0 0.0 ... 1 0 0.09885931558935361]
 [0.0 1.0 0.0 ... 0 0 0.03060836501901141]]

```

```

In [ ]: print(X_test)

```

```

[[1.0 0.0 0.0 ... 0 0 0.049429657794676805]
 [1.0 0.0 0.0 ... 0 1 0.09885931558935361]
 [0.0 1.0 0.0 ... 0 0 0.029863878326996197]
 ...
 [0.0 1.0 1.0 ... 0 0 0.515715969581749]
 [0.0 1.0 0.0 ... 0 0 0.03060836501901141]
 [0.0 1.0 0.0 ... 0 0 0.039923954372623575]]

```

```

In [ ]: df_train = pd.DataFrame(X_train)
        df_train

```

Out []:

	0	1	2	3	4	5	6	7	8	9
0	0.0	1.0	0.0	0.0	1.0	3	0.296306	0	0	0.030608
1	0.0	1.0	0.0	0.0	1.0	3	0.233476	0	0	0.038672
2	0.0	1.0	0.0	1.0	0.0	3	0.296306	0	0	0.029468
3	0.0	1.0	0.0	0.0	1.0	3	0.296306	0	0	0.030608
4	0.0	1.0	1.0	0.0	0.0	1	0.723549	0	2	0.430703
...
706	1.0	0.0	1.0	0.0	0.0	1	0.472229	0	0	0.865114
707	0.0	1.0	0.0	1.0	0.0	3	0.296306	1	0	0.091825
708	0.0	1.0	1.0	0.0	0.0	3	0.321438	1	0	0.054959
709	0.0	1.0	0.0	0.0	1.0	2	0.547625	1	0	0.098859
710	0.0	1.0	0.0	0.0	1.0	3	0.258608	0	0	0.030608

711 rows × 10 columns

```

In [ ]: df_test = pd.DataFrame(X_test)
        df_test

```

Out[]:

	0	1	2	3	4	5	6	7	8	9
0	1.0	0.0	0.0	0.0	1.0	2	0.447097	0	0	0.04943
1	1.0	0.0	0.0	0.0	1.0	2	0.623021	0	1	0.098859
2	0.0	1.0	0.0	0.0	1.0	3	0.597889	0	0	0.029864
3	1.0	0.0	0.0	1.0	0.0	3	0.195778	0	0	0.029468
4	0.0	1.0	0.0	0.0	1.0	3	0.170646	5	2	0.178327
...
173	0.0	1.0	0.0	0.0	1.0	3	0.296306	0	0	0.029642
174	0.0	1.0	1.0	0.0	0.0	3	0.132948	0	0	0.071435
175	0.0	1.0	1.0	0.0	0.0	1	0.271174	0	0	0.515716
176	0.0	1.0	0.0	0.0	1.0	3	0.271174	0	0	0.030608
177	0.0	1.0	0.0	0.0	1.0	2	0.296306	0	0	0.039924

178 rows × 10 columns